

### 3. REVIEW OF PROPERTIES OF EIGENVALUES AND EIGENVECTORS

#### 3.1 EIGENVALUES AND EIGENVECTORS

We shall now review some basic facts from matrix theory.

Let  $A$  be an  $n \times n$  matrix. A scalar  $\alpha$  is called an eigenvalue of  $A$  if there exists a nonzero  $n \times 1$  vector  $x$  such that

$$Ax = \alpha x$$

Example:

Let

$$A = \begin{pmatrix} -9 & 4 & 4 \\ -8 & 3 & 4 \\ -16 & 8 & 7 \end{pmatrix}$$

$$\alpha = -1$$

Consider

$$x = \begin{pmatrix} 1 \\ 2 \\ 0 \end{pmatrix}.$$

We have

$$\begin{aligned} Ax &= \begin{pmatrix} -9 & 4 & 4 \\ -8 & 3 & 4 \\ -16 & 8 & 7 \end{pmatrix} \begin{pmatrix} 1 \\ 2 \\ 0 \end{pmatrix} = \begin{pmatrix} -1 \\ -2 \\ 0 \end{pmatrix} = -1 \begin{pmatrix} 1 \\ 2 \\ 0 \end{pmatrix} \\ &= (-1)x = \alpha x \end{aligned}$$

Hence  $\alpha = -1$  is such that there exists a nonzero vector  $x$  such that  $Ax = \alpha x$ . Thus  $\alpha$  is an eigenvalue of  $A$ .

Similarly, if we take  $\alpha = 3$ ,  $x = \begin{pmatrix} 1 \\ 1 \\ 2 \end{pmatrix}$  we find that

$Ax = \alpha x$ . Thus,  $\alpha = 3$  is also an eigenvalue of  $A$ .

Let  $\alpha$  be an eigenvalue of  $A$ . Then any nonzero  $x$  such that  $Ax = \alpha x$  is called an eigenvector of  $A$ .

Let  $\alpha$  be an eigenvalue of  $A$ . Let,

$$W_\alpha = \{x \in C^n : Ax = \alpha x\}$$

Then we have the following properties of  $W_\alpha$  :

(i)  $W_\alpha$  is nonempty, since the zero vector, (which we denote by  $\theta$ ), is in  $W_\alpha$ , that is,  $A\theta_n = \theta_n = \alpha\theta_n$ .

(ii)  $x, y \in W_\alpha \Rightarrow Ax = \alpha x, Ay = \alpha y$   
 $\Rightarrow A(x + y) = \alpha(x + y)$   
 $\Rightarrow x + y \in W_\alpha$

(iii) For any constant k, we have

$kAx = k\alpha x = \alpha(kx)$   
 $\Rightarrow A(kx) = \alpha(kx)$   
 $\Rightarrow kx \in W_\alpha$

Thus  $W_\alpha$  is a subspace of  $C^n$ . This is called the characteristic subspace or the eigensubspace corresponding to the eigenvalue  $\alpha$ .

Example: Consider the A in the example on page 81. We have seen that  $\alpha = -1$  is an eigenvalue of A. What is  $W_{(-1)}$ , the eigensubspace corresponding to  $-1$ ?

We want to find all x such that

$$Ax = -x, \text{ that is,} \\ (A+I)x = \theta, \text{ that is,}$$

we want to find all solutions of the homogeneous system  $Mx = \theta$  ; where

$$M = A + I = \begin{pmatrix} -8 & 4 & 4 \\ -8 & 4 & 4 \\ -16 & 8 & 8 \end{pmatrix}$$

We now can use our row reduction to find the general solution of the system.

$$M \xrightarrow[\begin{matrix} R_2 - R_1 \\ R_3 - 2R_1 \end{matrix}]{\begin{matrix} R_2 - R_1 \\ R_3 - 2R_1 \end{matrix}} \begin{pmatrix} -8 & 4 & 4 \\ 0 & 0 & 0 \\ 0 & 0 & 0 \end{pmatrix} \xrightarrow{-\frac{1}{8}R_1} \begin{pmatrix} 1 & -\frac{1}{2} & -\frac{1}{2} \\ 0 & 0 & 0 \\ 0 & 0 & 0 \end{pmatrix}$$

Thus,  $x_1 = \frac{1}{2}x_2 + \frac{1}{2}x_3$

Thus the general solution of  $(A+I)x = \theta$  is

$$\begin{pmatrix} \frac{1}{2}x_2 + \frac{1}{2}x_3 \\ x_2 \\ x_3 \end{pmatrix} = \frac{1}{2}x_2 \begin{pmatrix} 1 \\ 2 \\ 0 \end{pmatrix} + \frac{1}{2}x_3 \begin{pmatrix} 1 \\ 0 \\ 2 \end{pmatrix} \\ = A_1 \begin{pmatrix} 1 \\ 2 \\ 0 \end{pmatrix} + A_2 \begin{pmatrix} 1 \\ 0 \\ 2 \end{pmatrix}$$

where  $A_1$  and  $A_2$  are arbitrary constants.

Thus consists of all vectors of the form

$$A_1 \begin{pmatrix} 1 \\ 2 \\ 0 \end{pmatrix} + A_2 \begin{pmatrix} 1 \\ 0 \\ 2 \end{pmatrix}.$$

Note: The vectors  $\begin{pmatrix} 1 \\ 2 \\ 0 \end{pmatrix}, \begin{pmatrix} 1 \\ 0 \\ 2 \end{pmatrix}$  form a basis for  $\omega_{-1}$  and therefore

$$\dim W_{(-1)} = 2.$$

What is  $W_{(3)}$  the eigensubspace corresponding to the eigenvalue 3 for the above matrix?

We need to find all solutions of  $Ax = 3x$ ,

$$\text{i.e., } Ax - 3x = \theta$$

$$\text{i.e., } Nx = \theta$$

where

$$N = A - 3I = \begin{pmatrix} -12 & 4 & 4 \\ -8 & 0 & 4 \\ -16 & 8 & 4 \end{pmatrix}$$

Again we use row reduction

$$N \xrightarrow{R_1 - \frac{2}{3}R_3 \text{ and } R_1 - \frac{4}{3}R_2} \begin{pmatrix} -12 & 4 & 4 \\ 0 & -\frac{8}{3} & \frac{4}{3} \\ 0 & \frac{8}{3} & -\frac{4}{3} \end{pmatrix} \xrightarrow{R_1 + R_3} \begin{pmatrix} -12 & 4 & 4 \\ 0 & -\frac{8}{3} & \frac{4}{3} \\ 0 & 0 & 0 \end{pmatrix}$$

$$\therefore 12x_1 = 4x_2 + 4x_3$$

$$\frac{8}{3}x_2 = \frac{4}{3}x_3 \quad \therefore x_3 = 2x_2$$

$$\therefore 12x_1 = 4x_2 + 8x_2 = 12x_2$$

$$\therefore x_2 = x_1$$

$$\therefore x_2 = x_1; x_3 = 2x_2 = 2x_1$$

$\therefore$  The general solution is

$$\begin{pmatrix} x_1 \\ x_1 \\ 2x_1 \end{pmatrix} = x_1 \begin{pmatrix} 1 \\ 1 \\ 2 \end{pmatrix}$$

Thus  $W_{(3)}$  consists of all vectors of the form

$$\kappa \begin{pmatrix} 1 \\ 1 \\ 2 \end{pmatrix}$$

Where  $\kappa$  is an arbitrary constant.

Note: The vector  $\begin{pmatrix} 1 \\ 1 \\ 2 \end{pmatrix}$  forms a basis for  $W_{(3)}$  and hence

$$\dim. W_{(3)} = 1.$$

Now when can a scalar  $\alpha$  be an eigenvalue of a matrix  $A$  of order  $n$ ? We shall now investigate this question. Suppose  $\alpha$  is an eigenvalue of  $A$ .

This  $\Rightarrow$  There is a nonzero vector  $x$  such that  $Ax = \alpha x$ .

$$\Rightarrow (A - \alpha I)x = \theta \text{ and } x \neq \theta$$

$$\Rightarrow \text{The system } (A - \alpha I)x = \theta \text{ has at least one nonzero solution.}$$

$$\Rightarrow \text{nullity } (A - \alpha I) \geq 1$$

$$\Rightarrow \text{rank } (A - \alpha I) < n$$

$$\Rightarrow (A - \alpha I) \text{ is singular}$$

$$\Rightarrow \det. (A - \alpha I) = 0$$

Thus,  $\alpha$  is an eigenvalue of  $A \Rightarrow \det. (A - \alpha I) = 0$ .

Conversely,  $\alpha$  is a scalar such that  $\det. (A - \alpha I) = 0$ .

This  $\Rightarrow (A - \alpha I)$  is singular

$$\Rightarrow \text{rank } (A - \alpha I) < n$$

$$\Rightarrow \text{nullity } (A - \alpha I) \geq 1$$

$$\Rightarrow \text{The system } (A - \alpha I)x = \theta \text{ has nonzero solution.}$$

$$\Rightarrow \alpha \text{ is an eigenvalue of } A.$$

Thus,  $\alpha$  is a scalar such that  $\det. (A - \alpha I) = 0 \Rightarrow \alpha$  is an eigenvalue.

Combining the two we get,

$\alpha$  is an eigenvalue of  $A$

$$\Leftrightarrow \det. (A - \alpha I) = 0$$

$$\Leftrightarrow \det. (\alpha I - A) = 0$$

Now let  $C(\lambda) = \det. (\lambda I - A)$

Thus we see that,

“The eigenvalues  $\lambda$  of a matrix  $A$  are precisely the roots of  $C(\lambda) = \det. (\lambda I - A)$ ”.

We have,

$$C(\lambda) = \begin{vmatrix} \lambda - a_{11} & -a_{12} & \dots & -a_{1n} \\ -a_{21} & \lambda - a_{22} & \dots & -a_{2n} \\ \dots & \dots & \dots & \dots \\ \dots & \dots & \dots & \dots \\ -a_{n1} & -a_{n2} & \dots & \lambda - a_{nn} \end{vmatrix}$$

$$= \lambda^n - (a_{11} + \dots + a_{nn})\lambda^{n-1} + \dots + (-1)^n \det A$$

Thus ;  $C(\lambda)$  is a polynomial of degree  $n$ . Note the 'leading' coefficient of  $C(\lambda)$  is one and hence  $C(\lambda)$  is a 'monic' polynomial of degree  $n$ . This is called CHARACTERISTIC POLYNOMIAL of  $A$ . The roots of the characteristic polynomial are the eigenvalues of  $A$ . The equation  $C(\lambda) = 0$  is called the characteristic equation.

Sum of the roots of  $C(\lambda) =$  Sum of the eigenvalues of  $A$

$$= a_{11} + \dots + a_{nn} ,$$

and this is called the TRACE of  $A$ .

Product of the roots of  $C(\lambda) =$  Product of the eigenvalues of  $A$

$$= \det A.$$

In our example in page 81 we have

$$A = \begin{pmatrix} -9 & 4 & 4 \\ -8 & 3 & 4 \\ -16 & 8 & 7 \end{pmatrix}$$

$$\therefore C(\lambda) = \det (\lambda I - A) = \begin{vmatrix} \lambda + 9 & -4 & -4 \\ 8 & \lambda - 3 & -4 \\ 16 & -8 & \lambda - 7 \end{vmatrix}$$

$$\xrightarrow{C_1 + C_2 + C_3} \begin{vmatrix} \lambda + 1 & -4 & -4 \\ \lambda + 1 & \lambda - 3 & -4 \\ \lambda + 1 & -8 & \lambda - 7 \end{vmatrix}$$

$$= (\lambda + 1) \begin{vmatrix} 1 & -4 & -4 \\ 1 & \lambda - 3 & -4 \\ 1 & -8 & \lambda - 7 \end{vmatrix}$$

$$\begin{aligned} \xrightarrow{\substack{R_2-R_1 \\ R_3-R_1}} &= (\lambda + 1) \begin{vmatrix} 1 & -4 & -4 \\ 0 & \lambda + 1 & 0 \\ 0 & -4 & \lambda - 3 \end{vmatrix} \\ &= (\lambda + 1)(\lambda + 1)(\lambda - 3) \\ &= (\lambda + 1)^2(\lambda - 3) \end{aligned}$$

Thus the characteristic polynomial is

$$C(\lambda) = (\lambda + 1)^2(\lambda - 3)$$

The eigenvalues are  $-1$  (repeated twice) and  $3$ .

$$\begin{aligned} \text{Sum of eigenvalues} &= (-1) + (-1) + 3 = 1 \\ &= \text{Trace } A = \text{Sum of diagonal entries.} \end{aligned}$$

$$\text{Product of eigenvalues} = (-1)(-1)(3) = 3 = \det. A.$$

Thus, if  $A$  is an  $n \times n$  matrix, we define the CHARACTERISTIC POLYNOMIAL as,

$$C(\lambda) = |\lambda I - A| \dots \dots \dots (1)$$

and observe that this is a monic polynomial of degree  $n$ . When we factorize this as,

$$C(\lambda) = (\lambda - \lambda_1)^{a_1} (\lambda - \lambda_2)^{a_2} \dots \dots (\lambda - \lambda_k)^{a_k} \dots \dots \dots (2)$$

where  $\lambda_1, \lambda_2, \dots, \lambda_k$  are the distinct roots; these distinct roots are the distinct eigenvalues of  $A$  and the multiplicities of these roots are called the algebraic multiplicities of these eigenvalues of  $A$ . Thus when  $C(\lambda)$  is as in (2), the distinct eigenvalues are  $\lambda_1, \lambda_2, \dots, \lambda_k$  and the algebraic multiplicities of these eigenvalues are respectively,  $a_1, a_2, \dots, a_k$ .

For the matrix in Example in page 81 we have found the characteristic polynomial on page 86 as

$$C(\lambda) = (\lambda + 1)^2(\lambda - 3)$$

Thus the distinct eigenvalues of this matrix are  $\lambda_1 = -1$  ; and  $\lambda_2 = 3$  and their algebraic multiplicities are respectively  $a_1 = 2$  ;  $a_2 = 1$ .

If  $\lambda_i$  is an eigenvalues of  $A$ , the eigen subspace corresponding to  $\lambda_i$  is  $W_{\lambda_i}$  and is defined as

$$W_{\lambda_i} = \{x \in C^n : Ax = \lambda_i x\}$$

The dimension of  $W_{\lambda_i}$  is called the GEOMETRIC MULTIPLICITY of the eigenvalue  $\lambda_i$  and is denoted by  $g_i$ .

Again for the matrix on page 81, we have found on pages 83 and 84 respectively that,  $\dim W_{(-1)} = 2$  ; and  $\dim W_{(3)} = 1$ . Thus the geometric multiplicities of the eigenvalues  $\lambda_1 = -1$  and  $\lambda_2 = 3$  are respectively  $g_1 = 2$  ;  $g_2 = 1$ . Notice that in this example it turns out that  $a_1 = g_1 = 2$  ; and  $a_2 = g_2 = 1$ . In general this may not be so. It can be shown that for any matrix A having C( $\lambda$ ) as in (2),

$$1 \leq g_i \leq a_i \quad ; \quad 1 \leq i \leq k \quad \dots \dots \dots (3)$$

i.e., for any eigenvalue  $\lambda_i$  of A, that is,

$1 \leq$  geometric multiplicity  $\leq$  algebraic multiplicity for any eigenvalue.

We shall now study the properties of the eigenvalues and eigenvectors of a matrix. We shall start with a preliminary remark on Lagrange Interpolation polynomials :

Let  $\alpha_1, \alpha_2, \dots, \alpha_s$  be s distinct scalars, (i.e.,  $\alpha_i \neq \alpha_j$  if  $i \neq j$ ). Consider,

$$p_i(\lambda) = \frac{(\lambda - \alpha_1)(\lambda - \alpha_2) \dots (\lambda - \alpha_{i-1})(\lambda - \alpha_{i+1}) \dots (\lambda - \alpha_s)}{(\alpha_i - \alpha_1)(\alpha_i - \alpha_2) \dots (\alpha_i - \alpha_{i-1})(\alpha_i - \alpha_{i+1}) \dots (\alpha_i - \alpha_s)}$$

$$= \prod_{\substack{1 \leq j \leq s \\ j \neq i}} \frac{(\lambda - \alpha_j)}{(\alpha_i - \alpha_j)} \quad \text{for } i = 1, 2, \dots, s \quad \dots \dots \dots (4)$$

Then  $p_i(\lambda)$  are all polynomials of degree s-1.

Further notice that  $p_i(\alpha_1) = \dots = p_i(\alpha_{i-1}) = p_i(\alpha_{i+1}) = \dots = p_i(\alpha_s) = 0$   
 $p_i(\alpha_i) = 1$

Thus  $p_i(\lambda)$  are all polynomials of degree s-1 such that,

$$p_i(\alpha_j) = \delta_{ij} = \begin{cases} 1 & \text{if } i = j \\ 0 & \text{if } i \neq j \end{cases} \quad \dots \dots \dots (5)$$

We call these the Lagrange Interpolation polynomials. If  $p(\lambda)$  is any polynomial of degree  $\leq s-1$  then it can be written as a linear combination of  $p_1(\lambda), p_2(\lambda), \dots, p_s(\lambda)$  as follows:

$$p(\lambda) = p(\alpha_1)p_1(\lambda) + p(\alpha_2)p_2(\lambda) + \dots + p(\alpha_s)p_s(\lambda) \dots (6)$$

$$= \sum_{i=1}^s p(\alpha_i)p_i(\lambda)$$

With this preliminary, we now proceed to study the properties of the eigenvalues and eigenvectors of an  $n \times n$  matrix  $A$ .

Let  $\lambda_1, \dots, \lambda_k$  be the distinct eigenvalues of  $A$ . Let  $\phi_1, \phi_2, \dots, \phi_k$  be eigenvectors corresponding to these eigenvalues respectively ; i.e.,  $\phi_i$  are nonzero vectors such that

$$A\phi_i = \lambda_i\phi_i, i=1,2,\dots,k \dots (6)$$

From (6) it follows that

$$A^2\phi_i = A(A\phi_i) = A(\lambda_i\phi_i) = \lambda_i A\phi_i = \lambda_i^2\phi_i$$

$$A^3\phi_i = A(A^2\phi_i) = A(\lambda_i^2\phi_i) = \lambda_i^2 A\phi_i = \lambda_i^3\phi_i$$

and by induction we get

$$A^m\phi_i = \lambda_i^m\phi_i \text{ for any integer } m \geq 0 \dots (7)$$

(We interpret  $A^0$  as  $I$ ).

Now,

$$p(\lambda) = a_0 + a_1\lambda + \dots + a_s\lambda^s$$

be any polynomial. We define  $p(A)$  as the matrix,

$$p(A) = a_0I + a_1A + \dots + a_sA^s$$

Now

$$p(A)\phi_i = (a_0I + a_1A + \dots + a_sA^s)\phi_i$$

$$= a_0\phi_i + a_1A\phi_i + \dots + a_sA^s\phi_i$$

$$= a_0\phi_i + a_1\lambda_i\phi_i + \dots + a_s\lambda_i^s\phi_i \text{ by (6)}$$

$$= (a_0 + a_1\lambda_i + \dots + a_s\lambda_i^s)\phi_i$$

$$= p(\lambda_i)\phi_i.$$

Thus,

If  $\lambda_i$  is any eigenvalue of  $A$  and  $\phi_i$  is an eigenvector corresponding to  $\lambda_i$ , then for any polynomial  $p(\lambda)$  we have  $p(A)\phi_i = p(\lambda_i)\phi_i$ .

PROPERTY I

Now we shall prove that the eigenvectors,  $\phi_1, \phi_2, \dots, \phi_k$  corresponding to the distinct eigenvalues  $\lambda_1, \lambda_2, \dots, \lambda_k$  of  $A$ , are linearly independent.

In order to establish this linear independence, we must show that

$$C_1\phi_1 + C_2\phi_2 + \dots + C_k\phi_k = \theta_n \Rightarrow C_1 = C_2 = \dots = C_k = 0 \dots (8)$$

Now if in (4) & (5) we take  $s = k$ ;  $\alpha_i = \lambda_i$ , ( $i=1,2,\dots,s$ ) then we get the Lagrange Interpolation polynomials as

$$p_i(\lambda) = \prod_{\substack{1 \leq j \leq k \\ j \neq i}} \frac{(\lambda - \lambda_j)}{(\lambda_i - \lambda_j)}; \quad i = 1, 2, \dots, k \quad \dots \dots \dots (9)$$

and

$$p_i(\lambda_j) = \delta_{ij} \quad \dots \dots \dots (10)$$

Now,

$$C_1\phi_1 + C_2\phi_2 + \dots + C_k\phi_k = \theta_n$$

For  $1 \leq i \leq k$ ,

$$p_i(A)[C_1\phi_1 + C_2\phi_2 + \dots + C_k\phi_k] = p_i(A)\theta_n = \theta_n$$

$$\Rightarrow C_1 p_i(A)\phi_1 + C_2 p_i(A)\phi_2 + \dots + C_k p_i(A)\phi_k = \theta_n$$

$$\Rightarrow C_1 p_i(\lambda_1)\phi_1 + C_2 p_i(\lambda_2)\phi_2 + \dots + C_k p_i(\lambda_k)\phi_k = \theta_n, \text{ (by property I on page 86)}$$

$$\Rightarrow C_i \phi_i = \theta; 1 \leq i \leq k; \text{ by (10)}$$

$$\Rightarrow C_i = 0; 1 \leq i \leq k \quad \text{since } \phi_i \text{ are nonzero vectors}$$

Thus

$C_1\phi_1 + C_2\phi_2 + \dots + C_k\phi_k = \theta_n \Rightarrow C_1 = C_2 = \dots = C_n = 0$  proving (8). Thus we have

Eigen vectors corresponding to distinct eigenvalues of A are linearly independent.

PROPERTY II